

Estimation question

Suppose that collection of  $N$  particles are each in state

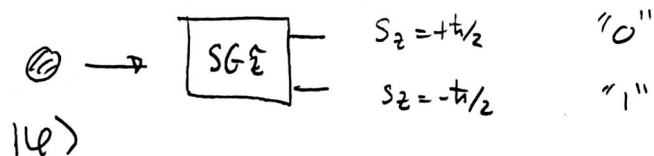
$$|\psi\rangle = \frac{1}{\sqrt{2}} |0\rangle + e^{i\phi} \frac{1}{\sqrt{2}} |1\rangle$$

We want to estimate  $\phi$ . We do so by subjecting each particle to the same measurement and recording measurement outcomes.

- Describe what direction  $SG_{\hat{n}}$  must be oriented to give one particular outcome with certainty.
- Suppose one measures  $SG_{\hat{n}}$  <sup>of each particle</sup>. Can one infer  $\phi$  from outcomes? If so how?
- Suppose one measures  $SG_{\hat{x}}$ . Can one infer  $\phi$  from outcomes? If so how.
- One of these is worse than other for learning  $\phi$ . For that one how is measurement direction related to  $\hat{n}$  (from part a).

## Measurement Operators

Suppose we measure  $SG_z$  or  $\{|0\rangle, |1\rangle\}$ . Then



We list

Outcome	Associated State	Prob
0 $\sim +\hbar/2$	$ 0\rangle$	$ \langle 0 \psi\rangle ^2$
1 $\sim -\hbar/2$	$ 1\rangle$	$ \langle 1 \psi\rangle ^2$

Then

$$\begin{aligned}
 \text{Prob}(0) &= |\langle 0|\psi\rangle|^2 = \langle 0|\psi\rangle^* \langle 0|\psi\rangle \\
 &= \langle \psi|0\rangle \langle 0|\psi\rangle
 \end{aligned}$$

$\langle 0|\psi\rangle^\dagger = (\langle 0|\psi\rangle)^*$   
 $\langle \psi|0\rangle$

We can extract the middle quantity  $\langle 0|0\rangle$  that refers to a measurement outcome. Then

$$\text{Prob}(0) = \underbrace{\langle \psi|}_{\text{bra}} \underbrace{(|0\rangle\langle 0|)}_{\text{measurement operator}} \underbrace{|\psi\rangle}_{\text{ket}}$$

What is the measurement operator?

$$|0\rangle\langle 0|$$

$$|0\rangle \rightarrow \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\langle 0| \rightarrow (1 \ 0)$$

$$|0\rangle\langle 0| \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$|\psi\rangle = a_0|0\rangle + a_1|1\rangle$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} = \begin{pmatrix} a_0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} = \begin{pmatrix} a_0 \\ 0 \end{pmatrix}$$

It can be regarded as a matrix. So define

$$\hat{P}_0 := |0\rangle\langle 0| \equiv \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\hat{P}_1 := |1\rangle\langle 1| \equiv \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

meas:

outcome	op	prob.
0	$\hat{P}_0$	$\langle \psi   \hat{P}_0   \psi \rangle$
1	$\hat{P}_1$	$\langle \psi   \hat{P}_1   \psi \rangle$

Then

$$\text{Prob}(0) = \langle \psi | \hat{P}_0 | \psi \rangle$$

$$\text{Prob}(1) = \langle \psi | \hat{P}_1 | \psi \rangle$$

Note that

$$\hat{P}_0 |\psi\rangle = |0\rangle\langle 0 | \psi \rangle = |0\rangle \underbrace{\langle 0 | \psi \rangle}_{\text{number}}$$

state.

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} = \begin{pmatrix} a_0 \\ 0 \end{pmatrix}$$

So  $\hat{P}_0$  maps a ket  $\rightarrow$  ket.

Exercise For each of the following states use the measurement operators to determine

a) the states  $\hat{P}_0|\psi\rangle$   
 $\hat{P}_1|\psi\rangle$ .

b) the probabilities of outcomes 0, 1.

States:

$$1) |\psi\rangle = |0\rangle$$

$$2) |\psi\rangle = \frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle$$

$$3) |\psi\rangle = \frac{1}{\sqrt{2}}|0\rangle - \frac{1}{\sqrt{2}}|1\rangle$$

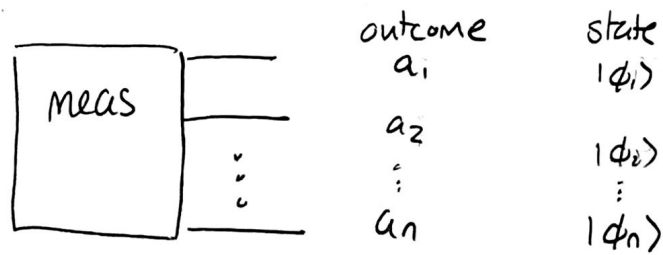
$$4) |\psi\rangle = \frac{1}{\sqrt{2}}|0\rangle + \frac{i}{\sqrt{2}}|1\rangle$$

$$5) |\psi\rangle = \frac{3}{5}|0\rangle + \frac{4i}{5}|1\rangle$$

$$\begin{pmatrix} \cdot & \cdot \end{pmatrix} \begin{pmatrix} \cdot & \cdot \\ \cdot & \cdot \end{pmatrix} \begin{pmatrix} \cdot \\ \cdot \end{pmatrix}$$

We can construct measurement operators for any measurement.

If



the measurement operators are:

$$\hat{P}_1 = |\phi_1\rangle\langle\phi_1|$$

$$\hat{P}_2 = |\phi_2\rangle\langle\phi_2|$$

$\vdots$

and  $\text{Prob}(\text{outcome } a_i) = \langle \overset{\text{state}}{\psi} | \hat{P}_i | \overset{\text{state in}}{\psi} \rangle$

Exercise a) Construct measurement operators for the measurement  $SG_y$ . Call these

$$\hat{P}_+ \leadsto S_y = +\hbar/2$$

$$\hat{P}_- \leadsto S_y = -\hbar/2$$

b) For the following pre-measurement states; determine probabilities:

1)  $|0\rangle$

2)  $\frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle$

3)  $\frac{1}{\sqrt{2}}|0\rangle + \frac{i}{\sqrt{2}}|1\rangle$

4)  $\frac{4}{5}|0\rangle + \frac{i3}{5}|1\rangle$

These

Measurement operators satisfy:

$$1) \hat{P}_i \hat{P}_i = \hat{P}_i$$

$$2) \hat{P}_i \hat{P}_j = 0 \quad i \neq j$$

$$3) \sum_{\text{all } i} \hat{P}_i = \hat{I} \rightarrow \text{identity} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

4) measure operators must be positive.

Exercise Check 1-3 for

a) operators for  $SG_z$   $\{|0\rangle, |1\rangle\}$

b) " "  $SG_y$

## Estimation (classical)

Coin toss.

$$\begin{aligned}\text{prob(heads)} &= p_H = p \\ \text{prob(tails)} &= p_T = 1-p\end{aligned}$$

- Want to estimate  $p$ .
- Flip coin  $N$  times. Count  $n_H$  = number heads  
 $n_T$  = " tails.

Estimate for  $p$  be?

$$p_H = \frac{n_H}{N} = p_{\text{estimate}} \stackrel{?}{=} p$$

- If repeat expt with  $N$  runs. — multiple times
  - get different  $n_H$
  - get different  $p_{\text{estimate}}$ .
- Can check if on average  $p_{\text{est}} = p$
- Want minimal fluctuations in  $p_{\text{est}}$ .
- Can calculate mean of  $p_{\text{estimate}}$ . Call this  $\bar{p}_{\text{estimate}}$

$$\bar{p}_{\text{estimate}} = \frac{1}{N} \bar{n}_H$$

- Can also calculate variance =  $\langle (p_{\text{est}} - \bar{p}_{\text{est}})^2 \rangle$  ← average  
std dev  
 $= \overline{p_{\text{est}}^2} - (\bar{p}_{\text{est}})^2$

Need

$$\overline{p_{\text{est}}^2} = \overline{\left( \frac{n_H^2}{N^2} \right)} = \frac{1}{N^2} \overline{n_H^2}$$

$$\overline{p_{\text{est}}^2} = \frac{1}{N^2} \overline{n_H^2}$$

Need  $\frac{\bar{n}_H}{n_H^2}$

$$\bar{n}_H = \sum_{n_H=0}^N n_H \text{prob}(n_H)$$

$$\bar{n}_H^2 = \sum_{n_H=0}^N n_H^2 \text{prob}(n_H)$$

Then

$$\text{prob}(n_H) = \binom{N}{n_H} p^{n_H} (1-p)^{N-n_H}$$

$$\sim \frac{N!}{n_H! (N-n_H)!}$$

$$\bar{n}_H = \sum_{n_H=0}^N n_H \binom{N}{n_H} p^{n_H} (1-p)^{N-n_H}$$

Binomial theorem:

$$(a+b)^N = \sum_{k=0}^N \binom{N}{k} a^k b^{N-k}$$

check

$$\sum_{n_H=0}^N \text{prob}(n_H) = \sum_{n_H=0}^N \binom{N}{n_H} p^{n_H} (1-p)^{N-n_H} = [p + (1-p)]^N = 1$$

$$\sum_{k=0}^N \binom{N}{k} a^k \frac{b^{N-k}}{(1-a)^{N-k}} = Na$$

$$\sum_{k=0}^N \binom{N}{k} a^k \frac{b^{N-k}}{(1-a)^{N-k}} k = Na[1 + a(N-1)]$$



Task:

\* Determine  $\bar{p}_{est}$

$$\hat{p} = p$$

\* Determine Variance ( $p_{est}$ )

small as possible

$$\hat{p} \sqrt{\frac{p(1-p)}{N}}$$

Was this the best one could do? Could do other things with n.H. Best requires

$$\bar{p}_{est} = \hat{p}$$

$$\text{variance}(p_{est}) = \text{smallest}$$

Theorem

$$\text{variance}(p_{est}) \geq \frac{1}{F_{N_{toss}}} \quad \text{Fisher info.}$$

Fisher info calculated via:

- probabilities for outcome of single coin toss and number of coin tosses (if these are independent)
- with N coin tosses

$$F_{N_{toss}} = N F_{\text{single coin toss.}}$$

calculated via probabilities for outcomes of single meas.

$$\sum_{\text{all possible outcomes}} \frac{1}{\text{prob}(\text{outcome})} \left( \frac{\partial \text{prob}(\text{outcome})}{\partial p} \right)^2 = F \quad \text{regardless of how one estimates}$$

\* Calculate  $F_{N_{toss}}$  + compare